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The Performance Characteristics of Some Reliability  
Growth Models

T. Jayachandran and L. R. Moore, III

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Department of Mathematics

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ROBERT R. FOSSUM  
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## SUMMARY

A reliability growth model is an analytical model that accounts for changes in reliability due to design changes and other corrective actions taken during the development and testing phases of a reliability program. This paper describes the results of a Monte Carlo study comparing the performance characteristics of four reliability growth models that have been proposed in the reliability literature.

## 1. INTRODUCTION

A reliability growth model is an analytical model that accounts for changes in reliability (usually an improvement) due to design changes and other corrective actions taken during the development and testing phases of a reliability program. A good reliability growth model would be useful in determining a test plan that will lead to the development of an acceptable product within a reasonable time period. Once a test plan is developed the model may be used to monitor the progress of the reliability program and if warranted, initiate changes such as accelerated testing. At any point in time during the test phase an idea as to how much additional testing would be required to achieve a reliability goal may be obtained using the model.

There are several reliability growth models available in the literature. Some of these models are probabilistic in nature i.e., the models do not allow for incorporation of information obtained during the test phase. The second type of models, generally known as statistical models, involve unknown parameters that can be estimated using available test data. This paper describes the results of a study comparing the performance for four statistical models. Two of these models are intended for use with time to failure data. The other two models use bernoulli type data; a "success" is taken to mean "the satisfactory operation for a specified period of time" and the specified time period is often referred to as "mission time".

A description of the reliability growth models selected for this study is given in Section 2. The Monte Carlo procedure used is described in Section 3 and Section 4 contains the methods of estimating the parameters of the different models. The results of the study are found in Section 5. Corcoran and Read [3] conducted a Monte Carlo study to compare several reliability growth models that includes two of the models considered in this paper. However, there are several differences in the methodology used in the two studies.

## 2. Description of the Reliability Growth Models.

Model I: Duane [4] developed this model based on his analysis of large quantities of data on various types of airborne equipment. If  $T$  is the accumulated test time on a given piece of equipment, and  $\lambda_T$  is the instantaneous failure rate at time  $T$  then, Duane's model states that

$$\lambda_T = \beta(1-\alpha)T^{-\alpha}$$

where  $\beta$  and  $\alpha$  are unknown parameters.

Model II: Suppose that (i) items are put on test sequentially and after each failure an appropriate corrective action is taken and (ii) the time to failure on the  $i^{\text{th}}$  test has an exponential distribution with parameter  $\lambda_i$ . Then, Weiss [5] assumes that the parameters  $\lambda_i$  satisfy the equation

$$\lambda_i = \frac{B + i}{Ai}$$

for some unknown parameters  $A$  and  $B$ .



Model III: Let

$p_r = P[\text{"mission success" after the } r^{\text{th}} \text{ failure has been observed and appropriate corrective action has been taken}]$

Chernoff and Woods [ 2 ] model  $p_r$  as an increasing function of  $r$  defined by

$$p_{r+1} = 1 - e^{-(\alpha+\beta r)}$$

Model IV: Wolman [ 6 ] developed this model assuming that

- (i) testing is conducted sequentially and each trial results in a "mission success" or a "mission failure"
- (ii) on each trial one of two types of failures can occur; the first type is an inherent or non-correctable failure and the second type of failure is a transient failure that can be eliminated by corrective action and
- (iii) whenever a test results in a "failure" an attempt is made to eliminate the failure mode and the attempt may result in a) either a complete elimination b) a partial elimination or c) no effect on the failure mode.

Let

$r$  = probability of an inherent failure, constant from test to test

$q$  = probability of a transient failure at the beginning of the test program

$\beta_i$  = factor by which the corrective action after the  $i^{\text{th}}$  failure, reduces the probability of a transient failure  $i-1$

$\alpha_i = \prod_{j=0}^i \beta_j$  the cumulative factor by which the initial

probability of a transient failure is reduced as a result of the corrective actions after each of the  $i$  failures and



$\beta_{n+1}(i)$  = probability of a "mission success" on the  
(n+1)st test if i failures are observed  
in the first n tests

The Wolman model assumes that

$$p_{n+1}(i) = 1 - r - q\alpha_i$$

Wolman points out that in many situations a good approximation is obtained by assuming that the improvement factors  $\beta_i$  are all equal to a constant value  $\beta$ . In this case, the model equation reduces to

$$p_{n+1}(i) = 1 - r - q\beta^i$$

The latter version with  $\beta = .2, .5$  and  $.8$  is used in this study.

### 3. Monte Carlo Procedure

In order to compare the growth models described in the previous section, five decreasing sequences of  $\lambda$  (failure rate) values are selected. Each sequence consists of 16 values ranging between .70 and .05; this represents a growth in reliability, assuming that all times are measured in mission units, from an initial value of .50 to a final value of .95 approximately. The growth patterns represented in these sequences vary from a very rapid growth to a very slow growth.

For each of the 16  $\lambda$ 's in a given sequence 10 exponential failure times  $t_{ij}$   $i=1,2,\dots,16$ ;  $j=1,2,\dots,10$  are generated. Since a reliability growth model is

to predict the  $\lambda$  values that are in practical situations unknown, the generated data is used to first estimate the  $\lambda$  values in a given sequence in two ways. In the first method the  $t_{ij}$ 's are treated as the observed exponential failure times and in the second case the  $t_{ij}$ 's are converted into bernoulli data by noting whether each  $t_{ij}$  exceeds 1 or not. This is done to study the effect of different data types on the prediction capabilities of the growth models. The estimated  $\lambda$  values are used to empirically determine the unknown parameters of the growth models using least squares techniques. To account for statistical variability the whole process described above is replicated 100 times resulting in 100 empirical growth curves for each of the  $\lambda$  sequences and each of the reliability growth models.

If  $\lambda_i$ ,  $i = 1, 2, \dots, 16$  are the failure rates in a given sequence 100 predicted values  $\lambda_i^*$  are obtained from the empirical curves for each of the models. The "goodness" of a reliability growth model is now measured using the following three measures:

$$M_1 = \sum_{i=1}^{16} \left| \frac{\lambda_i - \bar{\lambda}_i^*}{\lambda_i} \right|$$

$$M_2 = \sum_{i=1}^{16} \left[ \frac{\lambda_i - \bar{\lambda}_i^*}{\lambda_i} \right]^2$$

$$M_3 = \max_i \left| \frac{\lambda_i - \bar{\lambda}_i^*}{\lambda_i} \right|$$

where  $\bar{\lambda}_i^*$  is the average of the 100 predicted values  $\lambda_i^*$ . All three measures are relative measures i.e., each difference  $\lambda_i - \bar{\lambda}_i^*$  is divided

by the true value  $\lambda_i$  in computing the measures. The reason for choosing relative measures is that it would be important for a growth model to predict the small  $\lambda$  values at the end of a sequence rather well. In other words, these measures attach more weight to the differences  $\lambda_i - \bar{\lambda}_i^*$  at the end of each decreasing sequence.

#### 4. Estimation of Parameters

As indicated earlier the generated failure times  $t_{ij}$ ,  $j = 1, 2, \dots, 10$  are used to estimate the failure rates  $\lambda_i$  in two different ways. For the case where the  $t_{ij}$ 's are used directly the estimator of  $\lambda_i$  is the maximum likelihood estimator  $\hat{\lambda}_i = 1/\bar{t}_i$  where  $\bar{t}_i = \frac{1}{10} \sum_{j=1}^{10} t_{ij}$ . Since models III and IV describe the growth in  $p_i$  the probability of a mission success,  $\hat{\lambda}_i$  is converted into an estimator of  $p_i$  by noting that

$$\begin{aligned} p_i &= P[\text{mission success on } i^{\text{th}} \text{ trial}] \\ &= P[t_{ij} \geq 1] = e^{-\lambda_i} \end{aligned}$$

Thus, the maximum likelihood estimator of  $p_i$  is  $\hat{p}_i = e^{-\hat{\lambda}_i}$ . In the second case, the  $t_{ij}$ 's are transformed to bernoulli data and the  $\lambda_i$  and  $p_i$  are estimated from the transformed data as follows:

Let

$$x_{ij} = \begin{cases} 1 & \text{if } t_{ij} \geq 1 \\ 0 & \text{if } t_{ij} < 1 \end{cases}.$$

and let  $Y_i = \sum_{j=1}^k X_{ij}$ . Then,  $Y_i$  has a binomial distribution with parameters  $p_i$  and  $k$ . The maximum likelihood estimators of  $p_i$  and  $\lambda_i$  based on  $Y_i$  are  $\hat{p}_i = \frac{Y_i}{k}$  and  $\hat{\lambda}_i = -\ln \hat{p}_i$  respectively. The methods for obtaining the empirical growth curves and the predicted  $\lambda$  values from the growth curves are now described. Since these methods are the same for both sets of estimators  $\hat{\lambda}_i, \hat{p}_i$  and  $\tilde{\lambda}_i, \tilde{p}_i$  the symbols  $\tilde{\lambda}_i, \tilde{p}_i$  will be used to denote either of the two sets of estimators. Also, the symbols  $\lambda_i^*, p_i^*$  denote the  $\lambda_i, p_i$  values as predicted by the empirical growth curves.

For model I the equation describing the relationship between the failure rate  $\lambda_T$  and the total test time  $T$  is

$$\lambda_T = \beta(1-\alpha)T^{-\alpha}$$

or equivalently

$$\ln \lambda_T = \ln[\beta(1-\alpha)] - \alpha \ln T$$

The parameters  $\alpha$  and  $\beta$  are estimated by replacing  $\lambda_T$  with  $\tilde{\lambda}_i$  and  $T$  with  $T_i = \sum_{\ell=1}^i \sum_{j=1}^{10} t_{\ell j}$  in the latter equation and then applying standard regression methods. Once the least squares estimators  $\hat{\alpha}$  and  $\hat{\beta}$  are obtained, the predicted values  $\lambda_i^*$  of the failure rates  $\lambda_i$  are given by

$$\lambda_i^* = \hat{\beta}(1 - \hat{\alpha})T_i^{-\hat{\alpha}} \quad i = 1, 2, \dots, 16$$

The parameters  $A$  and  $B$  of model II are estimated using regression methods on the relationship

$$\tilde{\lambda}_i = \frac{B}{A} \frac{1}{i} + \frac{1}{A} \quad i = 1, 2, \dots, 16$$

and the predicted values are

$$\lambda_i^* = \frac{\hat{B} + i}{\hat{A}i} \quad i = 1, 2, \dots, 16$$

For model III the regression equation has the form

$$-\ln(1 - \tilde{p}_i) = \alpha + \beta(i-1) \quad i = 1, 2, 3, \dots, 16$$

and for model IV the equation is

$$1 - p_i = r + q\beta^{i-1} \quad i = 1, 2, \dots, 16 \quad \text{with } \beta \text{ known .}$$

The rest of the procedure is the same as for models I and II.

## 5. SIMULATION RESULTS

The results of the simulation are presented in tables 1.1 to 1.5. The five tables correspond to the five sets of 16  $\lambda$  values used in generating the failure times. These  $\lambda$  values are listed at the top of the tables. As indicated earlier the five sets chosen represent reliability growth ranging between a slow growth to a rapid growth. Also, the selection of the  $\lambda$ 's in these sets are made so as to generally follow the growth indicated by the models that are being tested. The entries in the tables are the observed values of the three measures  $M_1, M_2, M_3$  for each of the four models with three cases under model IV.

It may be observed from tables 1.1 to 1.5 that with time to failure data model I performed better than the others in all cases except one (table 1.4) where model III was best. For attributes data model IV with  $\beta = .8$  appears to outperform the other models. This result coincides with a similar conclusion by Bresenham [1].

In order to study the case where exactly one item is put on test at any time, the simulation was rerun generating only one failure time  $t_i$  instead of 10 failure times  $t_{ij}$  for each of the 16  $\lambda$ 's in a given sequence. There was a difference in the generation of attribute data

from the failure times. Let  $X_i$  denote the largest integer less than or equal to  $t_i$ , the time to failure on the  $i^{\text{th}}$  trial with failure rate  $\lambda_i$ ,  $i = 1, 2, \dots, 16$ . Since the  $t_i$  are assumed to be measured in mission units  $X_i$  is equivalent to the number of full missions completed before the occurrence of a failure. Thus,  $X_i$  has a geometric distribution with parameter  $p_i$ , the probability of a mission success on the  $i^{\text{th}}$  trial. The maximum likelihood estimator of  $p_i$  is  $\hat{p}_i = 1/(X_i + 1)$ . Except for this difference in the way the  $p_i$ 's are estimated the rest of the procedure is the same as before. The results of this second study are in tables 2.1 to 2.5. In this case, model III appears to do uniformly well with time to failure data and with attributes data model III and model IV with  $\beta = .8$  seems to perform better than the others.

On the basis of this study, the following general conclusions may be drawn:

- (1) with time to failure data and several items on test simultaneously, model I appears to predict reliability growth accurately
- (2) with attributes data and several items on test model IV with  $\beta = .8$  performs well
- (3) with time to failure data and one item on test model III is a good model to use and
- (4) with attributes data and one item on test either model III or model IV with  $\beta = .8$  may be used.

Graphs comparing true growth curves (curves down through the  $\lambda$  values used in the study) with the curves estimated by model I using time data and model IV using attribute data with  $\beta = .8$  are presented in Figures 1.1 to 1.5. Figures 2.1 to 2.5 compare the true growth curves with the curves estimated by model III using time and attribute data and model IV using attribute data with  $\beta = .8$ .

The computer programs used in the study are attached.



TABLE 1.1

True Values of Lambda: .702 .434 .320 .255 .213 .183 .161  
 .144 .130 .118 .109 .101 .0936 .0876 .0823 .0776

MODEL	TIME DATA			ATTRIBUTE DATA		
NUMBER	M1	M2	M3	M1	M2	M3
Model I	0.69	0.04	0.10	9.46	6.23	0.99
Model II	2.45	0.65	0.40	2.28	0.54	0.38
Model III	2.27	0.44	0.37	7.91	4.66	.778
Model IV $\beta = (.2)$	7.49	5.16	1.13	6.70	4.04	0.99
Model IV $\beta = (.5)$	4.52	2.13	0.79	4.09	1.57	0.67
Model IV $\beta = (.8)$	1.50	0.24	0.23	1.55	0.18	0.16

TABLE 1.2

True Values of Lambda: .700 .573 .508 .466 .435 .411 .392  
 .376 .362 .350 .340 .330 .322 .314 .307 .301

MODEL	TIME DATA			ATTRIBUTE DATA		
NUMBER	M1	M2	M3	M1	M2	M3
Model I	0.68	0.04	0.11	2.69	0.50	0.22
Model II	1.67	0.27	0.23	1.35	0.18	0.20
Model III	1.04	0.09	0.15	2.76	0.58	0.32
Model IV $\beta = (.2)$	2.67	0.58	0.33	2.21	0.40	0.25
Model IV $\beta = (.5)$	1.85	0.32	0.27	1.58	0.20	0.18
Model IV $\beta = (.8)$	1.09	0.08	0.11	0.39	0.01	0.06

TABLE 1.3

True Values of Lambda: .700 .353 .238 .180 .145 .122 .106  
 .0933 .0837 .0760 .0697 .0644 .0600 .0562 .0529 .0500

MODEL NUMBER	TIME DATA			ATTRIBUTE DATA		
	M1	M2	M3	M1	M2	M3
Model I	0.78	0.06	0.11	11.00	8.04	0.82
Model II	1.61	0.16	0.11	0.92	0.06	0.08
Model III	3.17	0.86	0.50	10.20	6.94	0.86
Model IV $\beta = (.2)$	8.42	6.57	1.26	7.33	4.88	1.08
Model IV $\beta = (.5)$	4.22	1.93	0.76	3.59	1.28	0.61
Model IV $\beta = (.8)$	3.87	1.16	0.43	3.81	1.13	0.44

TABLE 1.4

True Values of Lambda: .700 .564 .460 .379 .315 .263 .221  
 .186 .157 .132 .112 .0953 .0810 .0689 .0587 .0500

MODEL NUMBER	TIME DATA			ATTRIBUTE DATA		
	M1	M2	M3	M1	M2	M3
Model I	1.88	0.48	0.55	11.60	13.80	2.91
Model II	8.32	8.93	1.85	8.01	8.16	1.78
Model III	0.56	0.02	0.06	11.90	19.80	3.85
Model IV $\beta = (.2)$	14.10	24.30	3.04	12.90	20.10	2.77
Model IV $\beta = (.5)$	10.60	14.20	2.39	9.75	11.40	2.15
Model IV $\beta = (.8)$	2.45	0.92	0.65	1.95	0.47	0.48

TABLE 1.5

True Values of Lambda: .700 .456 .315 .227 .170 .132 .106  
 .0880 .0757 .0672 .0613 .0572 .0544 .0524 .0510 .0500

MODEL NUMBER	TIME DATA			ATTRIBUTE DATA		
	M1	M2	M3	M1	M2	M3
Model I	1.23	0.13	0.14	11.80	9.74	1.54
Model II	3.67	1.07	0.38	3.73	1.11	0.39
Model III	3.70	1.09	0.41	9.88	6.60	0.88
Model IV $\beta = (.2)$	11.80	12.10	1.47	10.80	10.20	1.34
Model IV $\beta = (.5)$	6.57	3.93	0.88	6.01	3.11	0.78
Model IV $\beta = (.8)$	3.68	1.12	0.47	3.47	1.07	0.53

TABLE 2.1

True Values of Lambda: .702 .434 .320 .255 .213 .183 .161  
 .144 .130 .118 .109 .101 .0936 .0876 .0823 .0776

MODEL NUMBER	TIME DATA			ATTRIBUTE DATA		
	M1	M2	M3	M1	M2	M3
Model I	19.20	43.10	5.44	5.71	6.15	2.21
Model II	13.20	11.30	10.80	30.7	83.8	4.83
Model III	9.63	6.37	0.76	4.56	1.68	.559
Model IV $\beta = (.2)$	40.50	297.00	15.40	16.50	85.00	6.26
Model IV $\beta = (.5)$	36.70	229.00	13.50	13.50	63.30	7.74
Model IV $\beta = (.8)$	23.00	35.00	35.00	3.88	1.37	0.57



TABLE 2.2

True Values of Lambda: .700 .573 .508 .466 .435 .411 .392  
 .376 .362 .350 .340 .330 .322 .314 .307 .301

MODEL NUMBER	TIME DATA			ATTRIBUTE DATA		
	M1	M2	M3	M1	M2	M3
Model I	18.60	38.40	5.02	3.73	3.51	1.78
Model II	118.00	877.00	8.35	8.68	14.10	3.15
Model III	6.36	2.70	0.49	2.16	.354	.240
Model IV $\beta = (.2)$	24.60	169.00	12.60	7.80	24.40	4.86
Model IV $\beta = (.5)$	24.40	164.00	12.40	7.27	23.10	4.75
Model IV $\beta = (.8)$	13.80	12.70	1.61	3.83	3.52	1.76

TABLE 2.3

True Values of Lambda: .700 .353 .238 .180 .145 .122 .106  
 .0933 .0837 .0760 .0697 .0644 .0600 .0562 .0529 .0500

MODEL	TIME DATA			ATTRIBUTE DATA		
NUMBER	M1	M2	M3	M1	M2	M3
Model I	19.20	41.20	5.20	5.47	6.17	2.26
Model II	127.00	1010.00	8.68	50.80	216.00	6.40
Model III	11.10	8.53	0.93	4.79	1.76	0.64
Model IV $\beta = (.2)$	45.80	303.00	14.60	15.90	60.50	7.30
Model IV $\beta = (.5)$	41.40	247.00	13.40	10.40	29.50	5.12
Model IV $\beta = (.8)$	25.90	44.50	2.12	3.54	.834	0.29

TABLE 2.4

True Values of Lambda: .700 .564 .460 .379 .315 .263 .221  
 .186 .157 .132 .112 .0953 .0810 .0689 .0587 .0500

MODEL NUMBER	TIME DATA			ATTRIBUTE DATA		
	M1	M2	M3	M1	M2	M3
Model I	21.00	70.00	7.49	8.13	11.90	2.91
Model II	182.00	2670.00	25.50	27.00	77.50	5.02
Model III	9.91	7.21	12.90	4.67	2.46	0.93
Model IV $\beta = (.2)$	49.20	433.00	17.40	25.90	167.00	11.90
Model IV $\beta = (.5)$	47.50	475.00	19.30	23.30	157.00	11.80
Model IV $\beta = (.8)$	28.40	81.60	5.72	8.54	10.10	1.91

TABLE 2.5

True Values of Lambda: .700 .456 .315 .227 .170 .132 .106  
 .0880 .0757 .0672 .0613 .0572 .0544 .0524 .0510 .0500

MODEL	TIME DATA			ATTRIBUTE DATA		
NUMBER	M1	M2	M3	M1	M2	M3
Model I	19.50	45.10	5.62	6.08	6.46	2.27
Model II	145.00	1370.00	11.10	48.50	205.00	6.34
Model III	10.90	9.22	1.11	5.50	2.21	0.59
Model IV $\beta = (.2)$	52.20	421.00	17.40	25.10	204.00	13.90
Model IV $\beta = (.5)$	46.30	353.00	16.50	18.20	124.00	10.90
Model IV $\beta = (.8)$	25.90	45.70	2.06	3.42	0.90	0.35

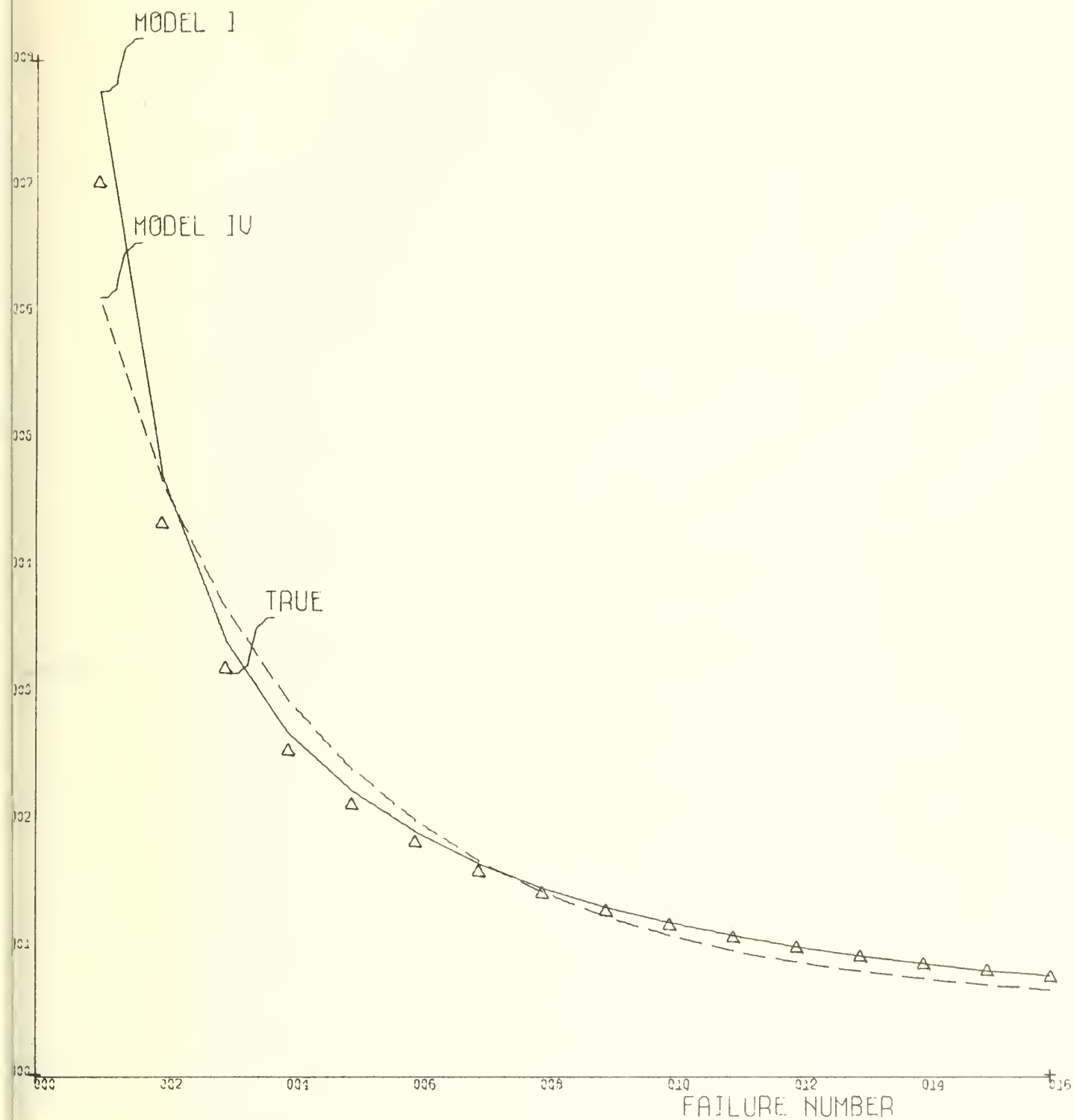


FIGURE 1.1 GRAPH OF PREDICTED FAILURE RATES FOR MODEL I, MODEL IV(.8), AND TRUE RATE

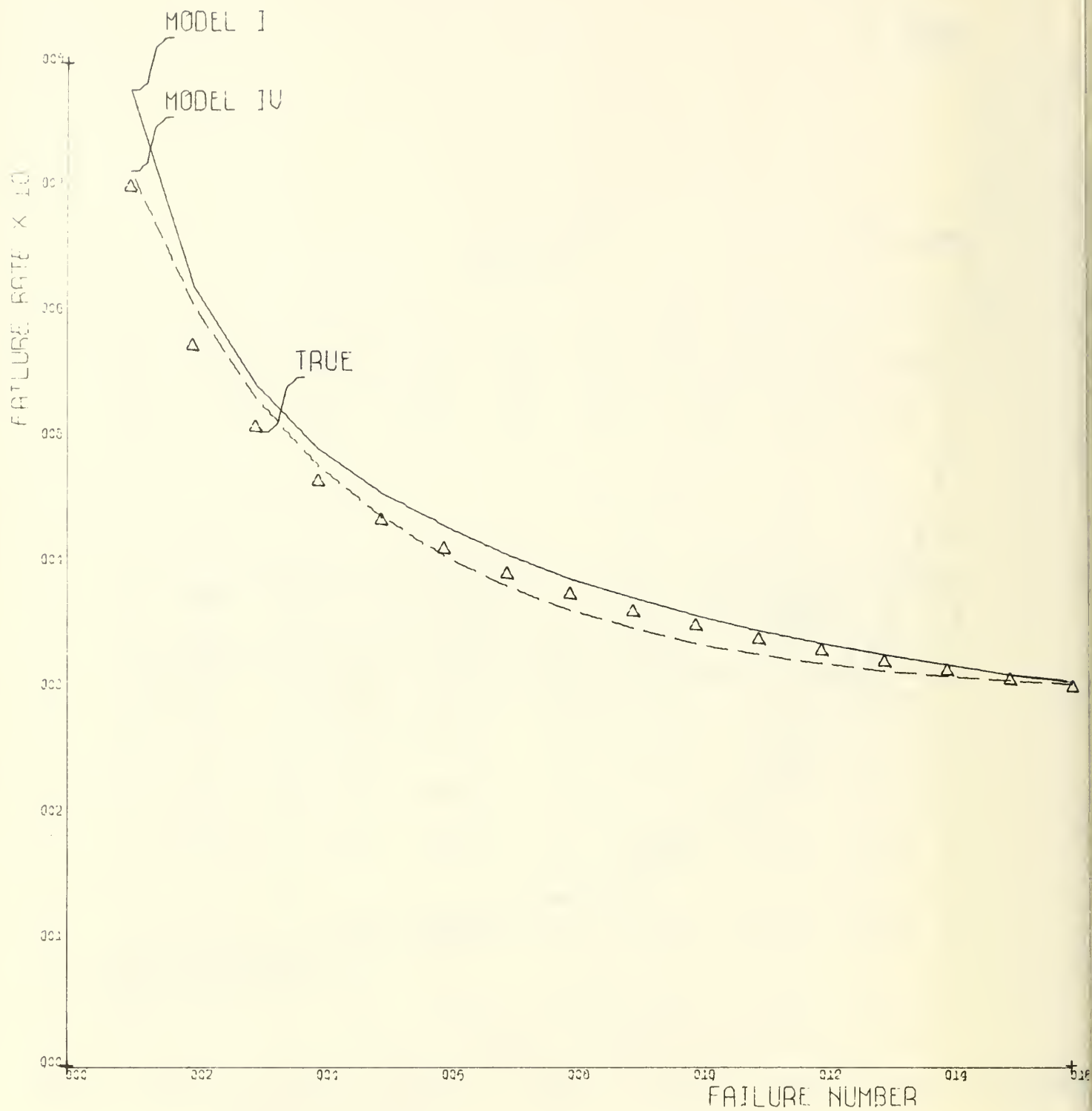


FIGURE 1.2 GRAPH OF PREDICTED FAILURE RATES FOR MODEL I, MODEL IV(.8), AND TRUE RATE



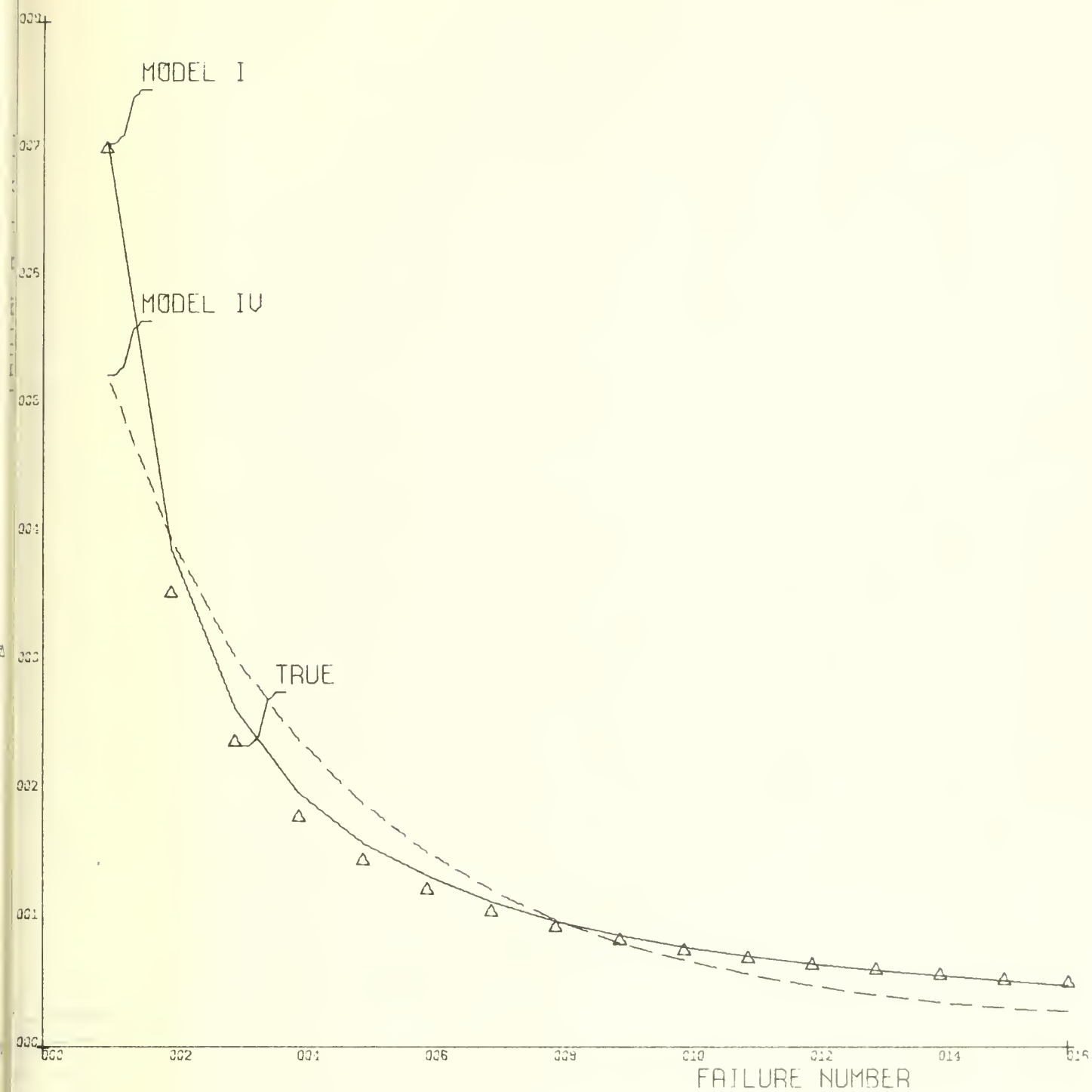


FIGURE 1.3 GRAPH OF PREDICTED FAILURE RATES FOR  
MODEL I, MODEL IV(.8), AND TRUE RATE

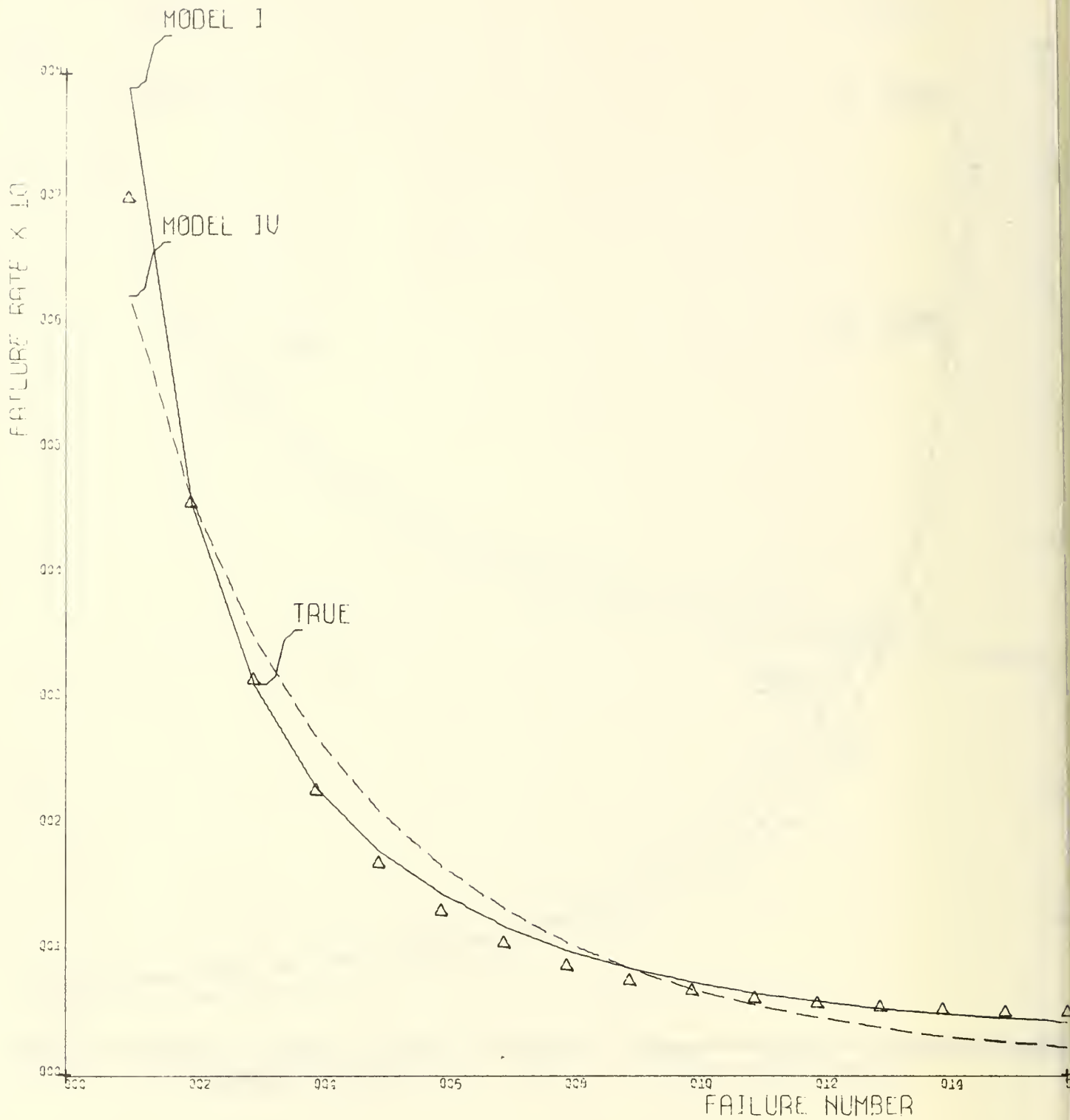


FIGURE 1.4 GRAPH OF PREDICTED FAILURE RATES F  
MODEL I, MODEL IV(.8), AND TRUE RA

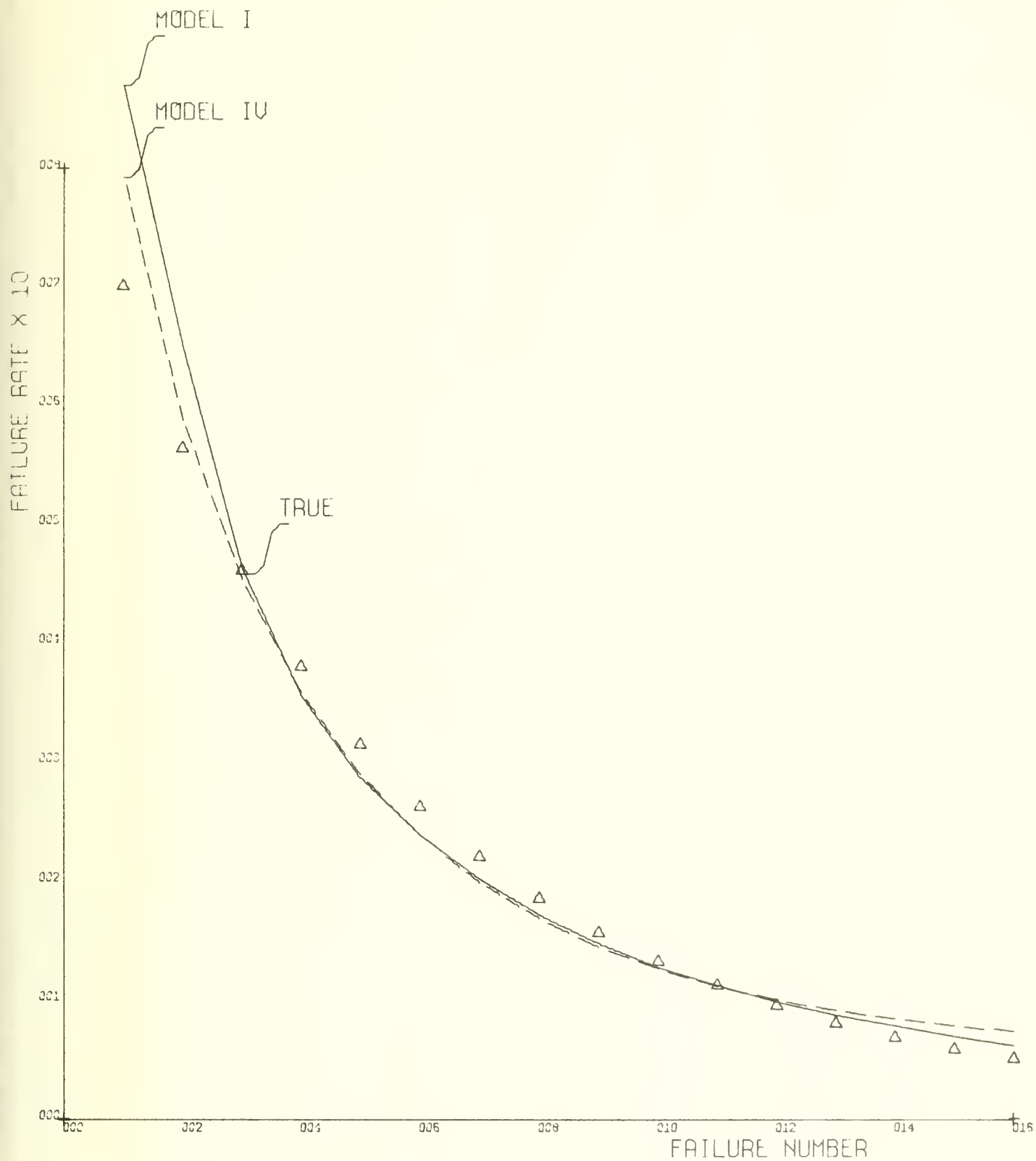


FIGURE 1.5 GRAPH OF PREDICTED FAILURE RATES FOR  
MODEL I, MODEL IV(.8), AND TRUE RATE

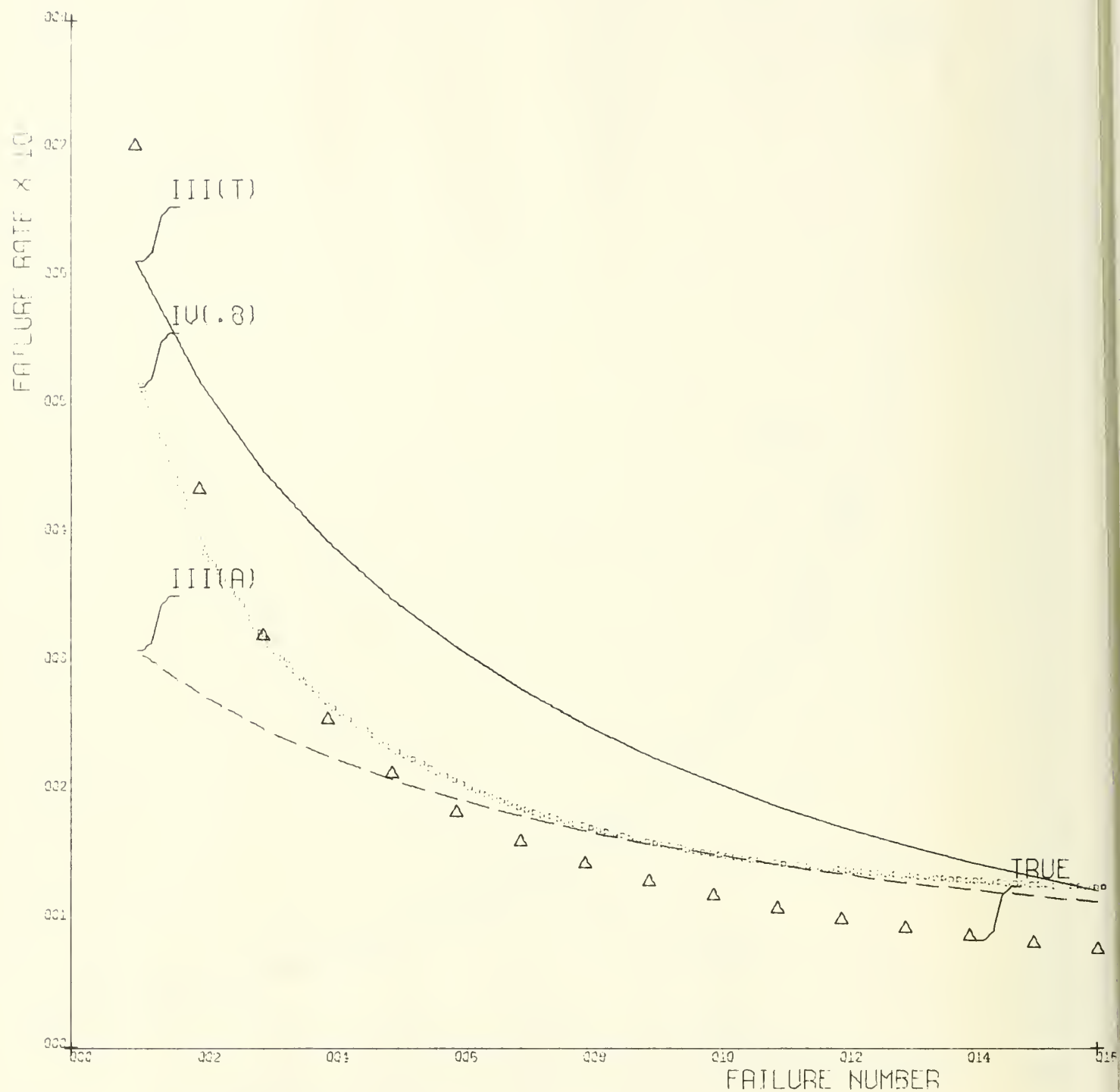


FIGURE 2.1 GRAPH OF FAILURE RATES PREDICTED BY MODELS III(T&A), IV(.8) AND TRUE RA

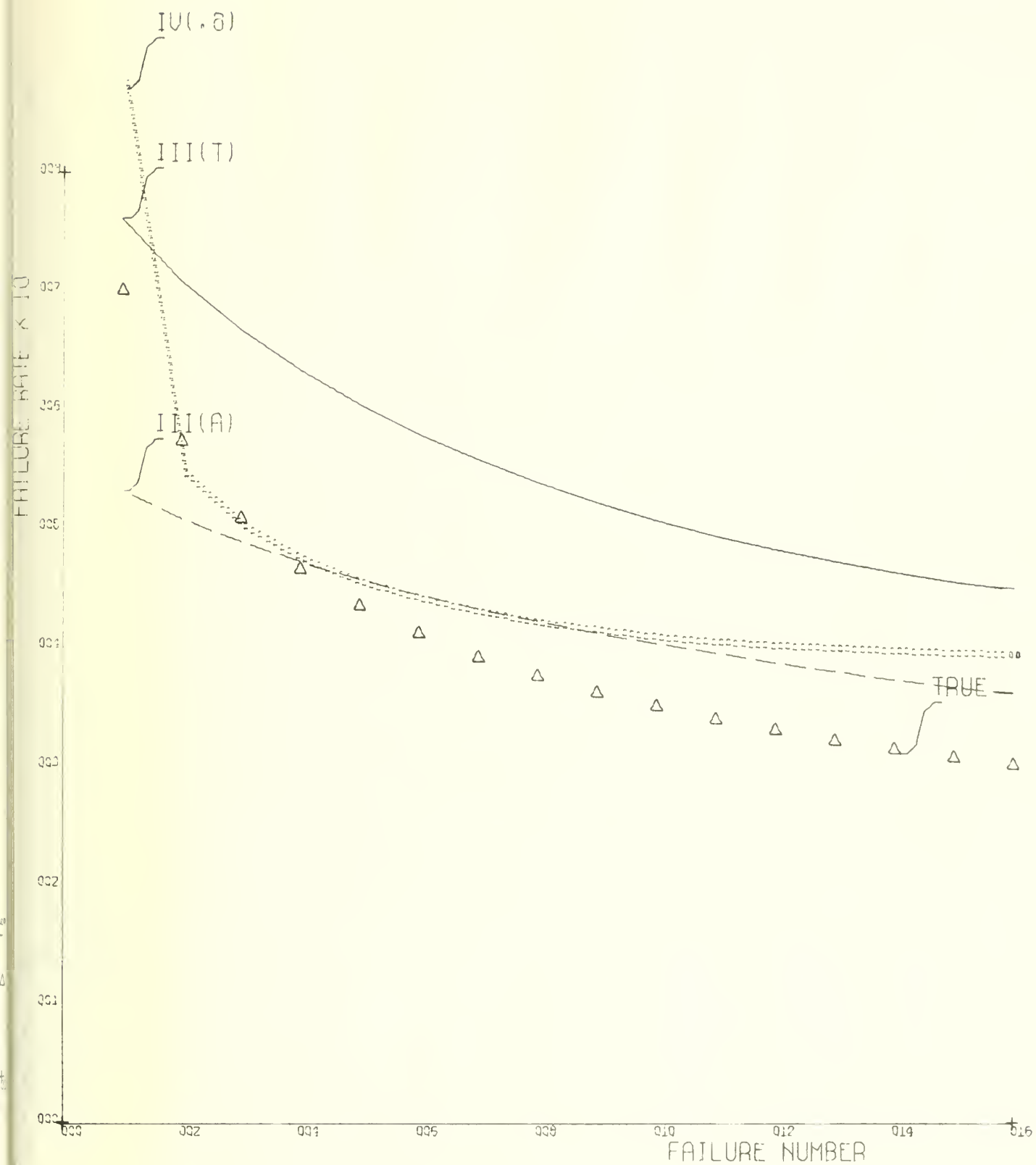


FIGURE 2.2 GRAPH OF FAILURE RATES PREDICTED BY MODELS  $III(T\&A)$ ,  $IV(.8)$  AND TRUE RATE

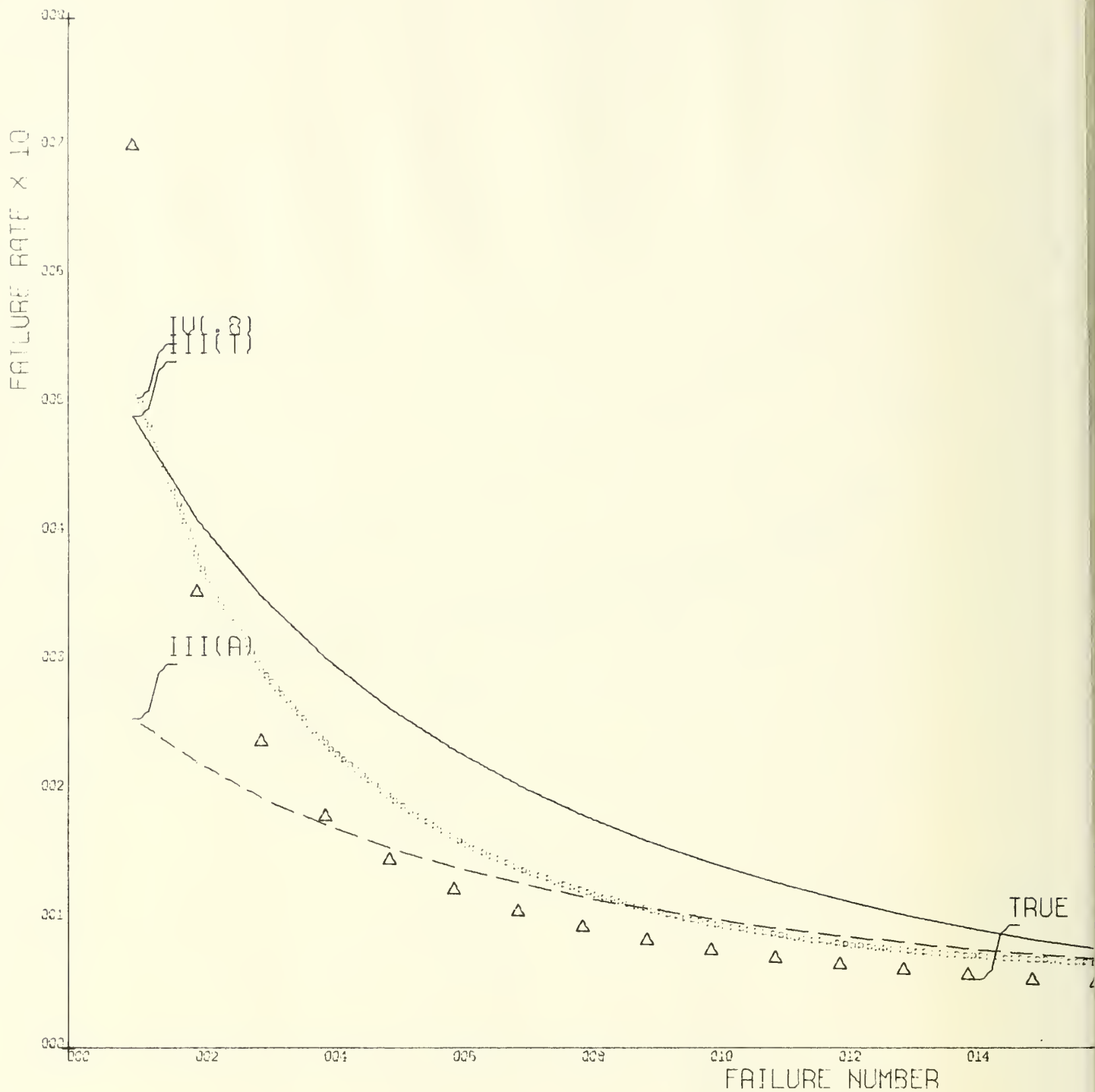


FIGURE 2.3 GRAPH OF FAILURE RATES PREDICTED BY MODELS III(T&A), IV(.8) AND TRUE F



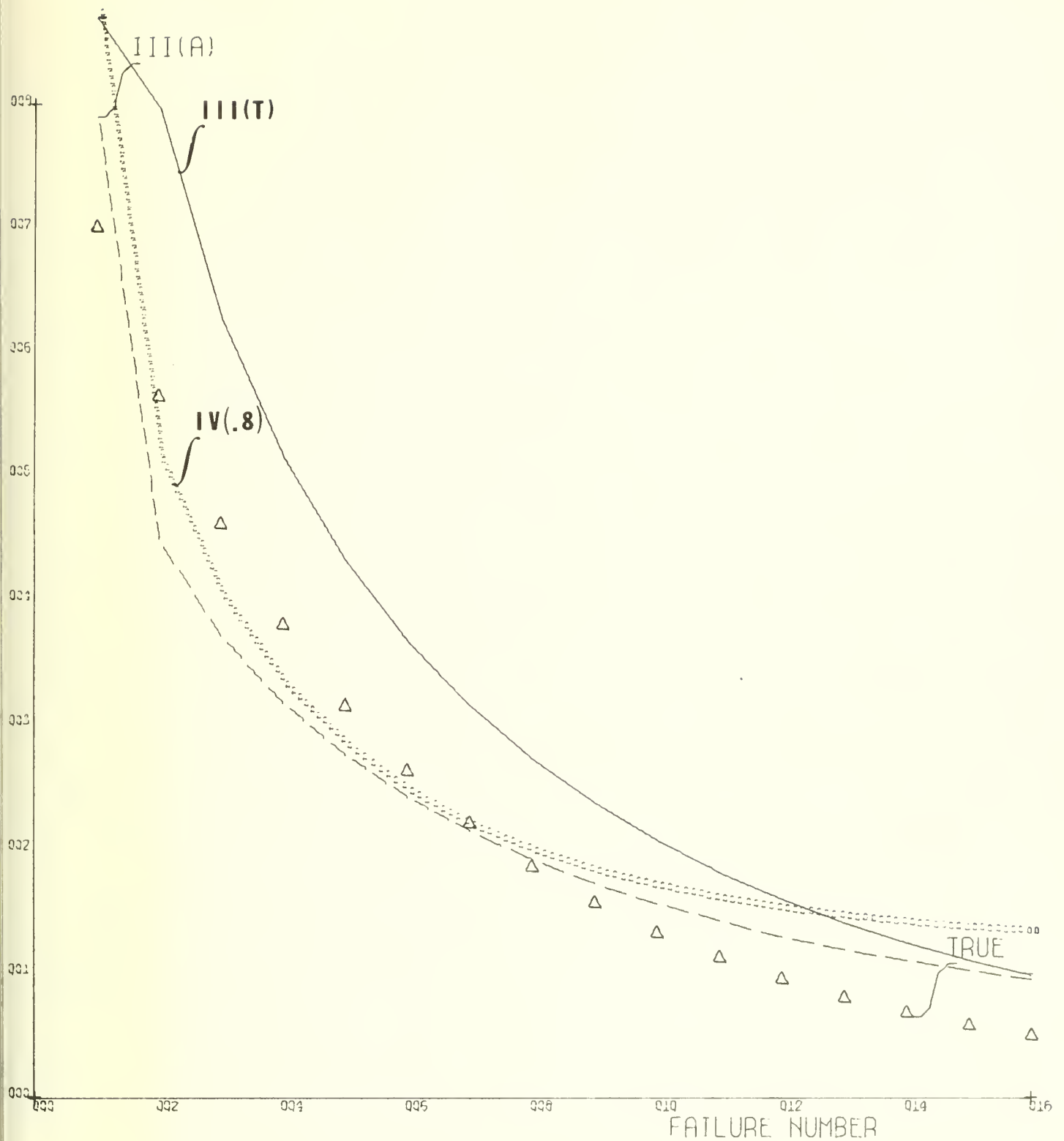


FIGURE 2.4 GRAPH OF FAILURE RATES PREDICTED BY MODELS III(T&A), IV(.8) AND TRUE RATE

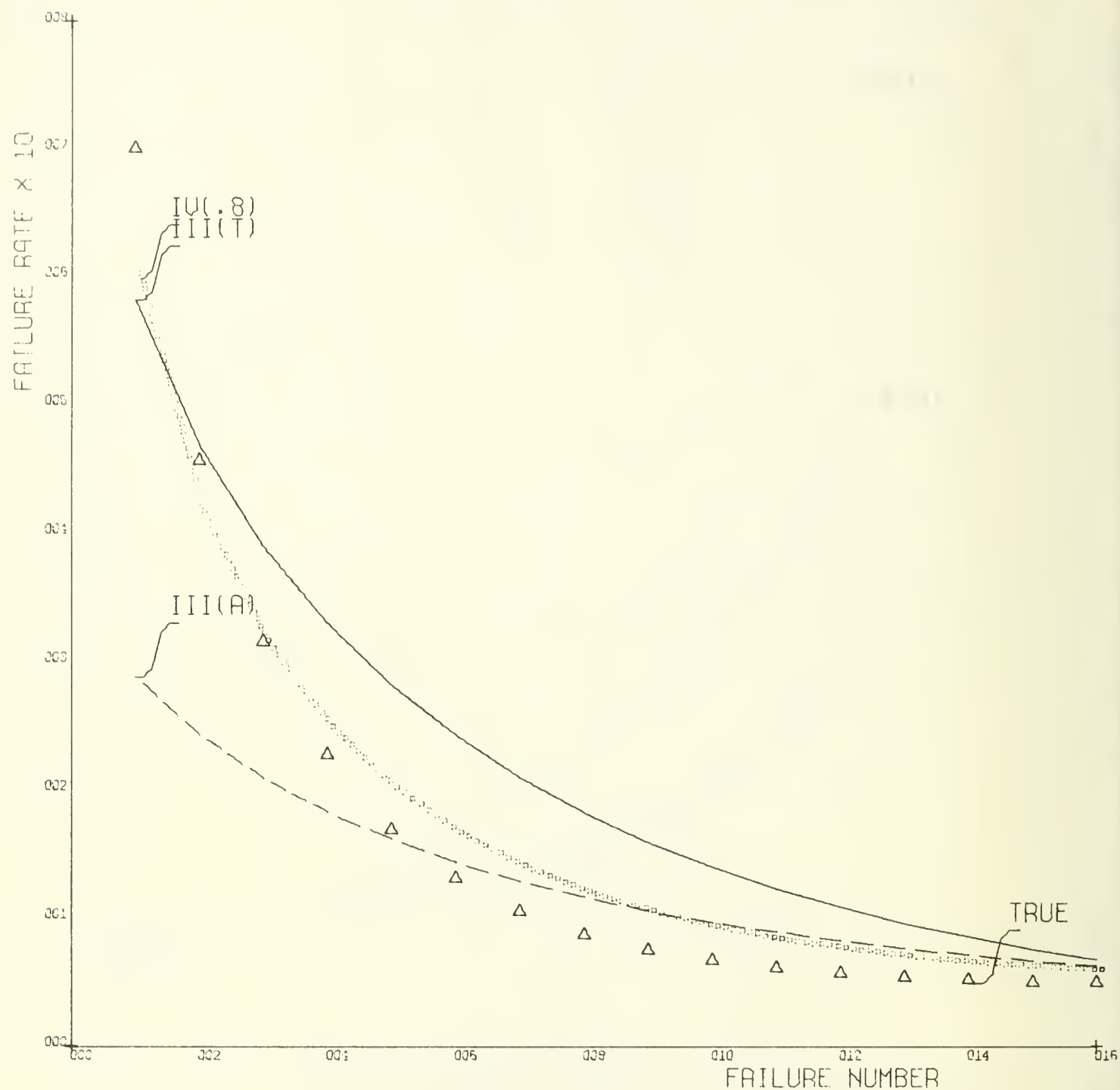


FIGURE 2.5 GRAPH OF FAILURE RATES PREDICTED BY MODELS III(T&A), IV(.8) AND TRUE RATE

### References

- [1] Bresenham, Jack E., "Reliability Growth Models," Stanford University Technical Report No. 74, August 1964.
- [2] Chernoff, Herman and Woods, W. Max, "Reliability Growth Models - - Analysis and Applications," CEIR, Inc., file memo dated February 26, 1962.
- [3] Corcoran, W. J. and Read, R. R., "Comparison of Some Reliability Growth Estimation and Prediction Schemes," UTC 2140-ITR Addendum, United Technology Center, Sunnyvale, California, June 1967.
- [4] Duane, J. T., "Learning Curve Approach to Reliability Monitoring," IEEE Transactions on Aerospace, Vol. 2, No. 2, 1964, pp. 563-566.
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- [6] Wolman, W., "Problems in System Reliability Analysis,: Statistical Theory of Reliability, University of Wisconsin, 1963.

```
// EXEC FORTCLG,REGION=200K
//FORT.SYSIN DD *

RELIABILITY GROWTH SIMULATION PROGRAM

***** THIS PROGRAM *****
SAMPLE DATA DECK FOLLOWS THIS PROGRAM *****
*****

INPUT
  DEVICE 5 "CARDS"
  COL 1-4
    KEY WORD
    FINISH => TERMINATES PROGRAM
    NEW => NEW SET OF TRUE RATES FOLLOW
    DATA IS INPUT VIA NAMELISTS LAMDA AND TIMES
    &LAMDA NUM=NN,NOFMT=FF &END

WHERE NN IS THE NUMBER OF LAMDAS
AND FF IS THE NUMBER OF FORMAT CARDS
TRUE LAMDA VALUES FOLLOW FORMAT CARDS

&TIMES NOREP=RR,ISEED=SSSSSSS &END

WHERE RR = NUMBER OF REPETITIONS AND
SSSSSSSS = THE SEED FOR THE SIMULATION

(THE NEXT 4 KEY WORDS CAUSE THE MODELS
INDICATED TO PREDICT THE TRUE LAMDAS)
GENERAL=> MODEL I
WEISS=> MODEL II
WOODS=> MODEL III
WOLMAN=> MODEL IV   COL 41 - 50 ARE FOR BETA

RESET => RESETS ALL PRINT OPTIONS TO INITIAL CONDITIONS
COL 41 - 50 = FIRST MEASURE PRINTED
COL 51 - 60 = NUMBER OF SIMULATIONS
COL 71 - 80 = OUTPUT DEVICE FOR MEASURES
COL 61 - 70 = LAST MEASURE PRINTED
PRINT=> PRINT OUT RESULTS TO DATE
A CARD WITH NO KEY WORD CAUSES THE PREDICTIONS
BASED ON THE RANDOM NUMBERS GENERATED
TO BE USED AS IF THEY WERE THE RESULT OF
ONE OF THE MODELS
```

CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC

OUTPUT  
DEVICE 6 "PRINTER"  
RESULTS OF EACH SIMULATION  
USUALLY DUMMIED FOR LARGE NUMBER OF SIMULATIONS  
  
DEVICE IN COL 71 - 80 OF RESET CARD  
FINAL RESULTS FOR MEASURES SPECIFIED  
  
DEVICE 9 "CARDS"  
PREDICTED LAMDAS ARE OUTPUT FOR PLOTTING

\*\*\*\*\*  
THIS PROGRAM CALCULATES AND OUTPUTS OTHER QUANTITIES  
THAN THE ONES MENTIONED ABOVE FOR FUTURE RESEARCH.  
\*\*\*\*\*

PROGRAMS CALLED OTHER THAN THOSE PRESENTED BELOW  
ARE IN THE NPS PROGRAM LIBRARY  
OF SPECIAL SIGNIFICANCE IS THE PROGRAM "LLRANDOM"  
WHICH GENERATES OUR RANDOM NUMBERS

REAL\*8 HOLDX(300)  
REAL\*8 HOLDX(300)  
DIMENSION FITTED(301)  
DIMENSION AMDA(301)  
REAL\*4 MODEL(10)  
DIMENSION PARM(4)  
REAL\*4 MEAS(8,6)  
REAL\*4 WOLM/'WOLM'/  
REAL\*4 NEWL/'NEWL'/  
REAL\*4 WEIS/'WEIS'/  
REAL\*4 MOOR/'MOOR'/  
REAL\*4 GENE/'GENE'/  
REAL\*4 FINI/'FINI'/  
REAL\*4 WOOD/'WOOD'/  
REAL\*4 RESE/'RESE'/,PRIN/'PRIN'/  
LOGICAL ON  
NOUT=6  
NSIM=1  
NMEAS=3  
CONTINUE

PUB000010  
PUB000020  
PUB000030  
PUB000040  
PUB000050  
PUB000060  
PUB000070

PUB000090

PUB000090

```

100 IF(NMEAS.LE.0)NMEAS=3
101 IF(MSEND.LE.0)MSEND=NMEAS
      IF(NSIM.LE.0)NSIM=1
      IF(NOUT.LE.0)NOUT=6
      READ(5,100,END=999)MODEL,PARM
      FORMAT(10A4,4F10.0)
      WRITE(6,101)MODEL
      FORMAT(11,/,/,/,10A4,/)
      IF(MODEL(1).NE.RESE)GO TO 4
      CALL RESET
      NMEAS=PARM(1)+.5
      NSIM=PARM(2)+.5
      NOUT=PARM(3)+.5
      MSEND=PARM(4)+.5
      GO TO 1
4     CONTINUE
      IF(MODEL(1).NE.PRIN)GO TO 5
      CALL DMPOUT(NOUT)
      GO TO 1
5     CONTINUE
      IF(MODEL(1).EQ.FINI)CALL FINISH(NOUT)
      IF(MODEL(1).NE.NEWL)GO TO 8
      CALL START(AMDA,NO,ISEED)
      GO TO 1
8     CONTINUE
      DO 7 K=1,300
      HOLDX(K)=0.0D+00
      HOLDXX(K)=0.0D+00
      CONTINUE
      ON=.TRUE.
      NEND=NO*6
      NSEED=IABS(ISEED)/2*2+1
      DO 6 K=1,NSIM
      CALL SMLT(NSSEED,FITTED,ON)
      IF(MODEL(1).EQ.WOOD)CALL WOODS(AMDA,NO,FITTED)
      IF(MODEL(1).EQ.GENE)CALL GE(AMDA,NO,FITTED)
      IF(MODEL(1).EQ.WEIS)CALL WEISS(AMDA,NO,FITTED)
      IF(MODEL(1).EQ.WOLM)CALL WOLMAN(AMDA,NO,PARM(1),FITTED)
      DO 9 LMN=1,NEND
      HOLDX(LMN)=FITTED(LMN)+HOLDX(LMN)
      HOLDXX(LMN)=HOLDXX(LMN)+FITTED(LMN)**2
      CONTINUE
9     IF(ON)WRITE(NOUT,888)MODEL,NSIM
      FORMAT(10,10A4,/,/,' SIMULATION WILL BE REPEATED',I4,' TIMES.:',
1/,' LAMDA ESTIMATES WILL NOT BE PRINTED FOR EACH SIMULATION.:',)
      ON=.FALSE.
      CONTINUE
      DO 10 K=1,NEND

```

PUB00100  
PUB00110  
PUB00120  
PUB00130

PUB00140





11	SUM=0.0D+00 DO 11 K=1,NUM SUM=SUM+ERROR(K,NOEST) CONTINUE XMEAS=SUM RETURN	PUB00530 PUB00540 PUB00550 PUB00560 PUB00570 PUB00580 PUB00590
20	CONTINUE SUM=0.0D+00 DO 21 K=1,NUM SUM=SUM+ABS(ERROR(K,NOEST)) CONTINUE XMEAS=SUM RETURN	PUB00600 PUB00610 PUB00620 PUB00630 PUB00640 PUB00650 PUB00660
21	CONTINUE SUM=0.0D+00 DO 31 K=1,NUM SUM=SUM+ERROR(K,NOEST)**2 CONTINUE XMEAS=SUM RETURN	PUB00670 PUB00680 PUB00690 PUB00700 PUB00710 PUB00720 PUB00730
30	CONTINUE SUM=0.0D+00 DO 41 K=1,NUM SUM=SUM+REL(K,NOEST) CONTINUE XMEAS=SUM RETURN	PUB00740 PUB00750 PUB00760 PUB00770 PUB00780 PUB00790 PUB00800
31	CONTINUE SUM=0.0D+00 DO 51 K=1,NUM SUM=SUM+ABS(REL(K,NOEST)) CONTINUE XMEAS=SUM RETURN	PUB00810 PUB00820 PUB00830 PUB00840 PUB00850 PUB00860 PUB00870
40	CONTINUE SUM=0.0D+00 DO 61 K=1,NUM SUM=SUM+REL(K,NOEST)**2 CONTINUE XMEAS=SUM RETURN	PUB00880 PUB00890 PUB00900 PUB00910 PUB00920 PUB00930 PUB00940
41	CONTINUE SUM=0.0D+00 DO 71 K=1,NUM IF(ABS(ERROR(K,NOEST)).GT.SUM)SUM=ABS(ERROR(K,NOEST)) CONTINUE XMEAS=SUM RETURN	PUB00950 PUB00960 PUB00970 PUB00980 PUB00990 PUB01000
50		
51		
60		
61		
70		
71		

```

80 CONTINUE
SUM=0.0D+00
DO 81 K=1,NUM
IF(ABS( REL(K,NJEST)).GT.SUM)SUM=ABS( REL(K,NJEST))
81 CONTINUE
XMEAS=SUM
RETURN
END
SUBROUTINE PLOTEM(A,NO,F,WORDS)
DIMENSION A(NO,1),F{NO,1},WORDS(1)
COMMON/WORK/ORDER(600)
DIMENSION TITLE(23)
DIMENSION ARRAY(50,2,7)
DATA TITLE/92HPLOT OF EST NO 1-6(CURVE 1-6) AND TRUE(CURVE 7) FOR
1 DIMENSION INUM(7)
NUM=50
NPLOT=7
DO 1 K=1,7
INUM(K)=NO
CONTINUE
DO 3 I=1,NO
DO 2 K=2,7
KM1=K-1
ARRAY(I,2,KM1)=F(I,KM1)
ARRAY(I,1,KM1)=I
CONTINUE
ARRAY(I,2,7)=A(I,1)
ARRAY(I,1,7)=I
CONTINUE
MINUS=-1
DO 5 L=1,10
LP13=L+13
TITLE(LP13)=WORDS(L)
CONTINUE
CALL PPLOT(NUM,NPLOT,INUM,ARRAY,5,5HLMADA,10,10HFAILURE NO,
1 MINUS,MINUS,92,TITLE,ORDER)
RETURN
END
SUBROUTINE START(AMDA,NO,IX)
NAMELIST /LAMDA/NUM,NOFMT
NAMELIST /TIMES/NOREP,ISEED
DIMENSION FITTED(1)
COMMON/WORK/FORM(600)
COMMON/EXPTIM/NOR,T(2500)
DIMENSION AMDA(1)
REAL*4 BLANK(3)/'(' ',' ',' ')/
LOGICAL ON

```

```

PUB01010
PUB01020
PUB01030
PUB01040
PUB01050
PUB01060
PUB01070
PUB01080
PUB01090
PUB01100
PUB01110
PUB01120
PUB01130
PUB01140
PUB01150
PUB01160
PUB01170
PUB01180
PUB01190
PUB01200
PUB01210
PUB01220
PUB01230
PUB01240
PUB01250
PUB01270
PUB01290
PUB01300
PUB01310
PUB01320
PUB01330
PUB01340
PUB01350
PUB01360
PUB01370
PUB01380
PUB01390
PUB01410
PUB01420
PUB01430
PUB01440
PUB01450
PUB01460
PUB01470

```

```

1      FORM(1)=BLANK(1)
      DO 1 K=2,599
      FORM(K)=BLANK(2)
      CONTINUE
      FORM(600)=BLANK(3)
      NUM=1
      NOFMT=1
      READ(5,LAMDA)
      NTOT=NOFMT*20
      READ(5,100)(FORM(I),I=2,NTOT)
      FORMAT(20A4)
      READ(5,FORM)(AMDA(I),I=1,NUM)
      ISEED=123456789
      NOREP=1
      READ(5,TIMES)
      NCR=NOREP
      NO=NUM
      IX=ISEED
      LDSEED=0
      RETURN
      ENTRY SMLT(NSEED,FITTED,ON)
      IX=NSEED
      IF(IX.EQ.LDSEED)GO TO 25
      LDSEED=IX
      N=NO*NOR
      CALL EXPON(IX ,T,N)
      INDX=0
      DO 2 K=1,NUM
      DO 2 J=1,NOREP
      INDX=INDX+1
      T(INDX)=T(INDX)/AMDA(K)
      CONTINUE
      CALL EST(AMDA,NO ,T,NOR)
      IF(ON)WRITE(6,101)NOR
      101  FORMAT( //,' LAMDA AND LAMDA ESTIMATES ',//,MLE(INTEGER DATA),',
      1, EST #1 => TRUE, EST #2 => MLE, EST #3 => MLE(INTEGER DATA),',
      2, EST #4 => RMLE, //,' EST #5 => RMLE(INTEGER DATA), EST #6 => AT
      3IBUTE DATA, //,' BASED ON ',12,' REPETITIONS',//)
      25  CONTINUE
      DO 12 J=1,6
      LL=J*NO
      L=LL-NO+1
      DO 11 K=L,LL
      FITTED(K)=AMDA(K)
      CONTINUE
      11  IF(IX.EQ.LDSEED)GO TO 12
      IF(.NOT.ON)GO TO 12
      WRITE(6,105)J,(AMDA(K),K=L,LL)
      PUB01480
      PUB01490
      PUB01500
      PUB01510
      PUB01520
      PUB01530
      PUB01540
      PUB01550
      PUB01560
      PUB01570
      PUB01580
      PUB01590
      PUB01600
      PUB01610
      PUB01620
      PUB01630
      PUB01640
      PUB01650
      PUB01660
      PUB01670
      PUB01680
      PUB01690
      PUB01700
      PUB01710
      PUB01740
      PUB01760
      PUB01770
      PUB01780
      PUB01790
      PUB01800
      PUB01810
      PUB01820
      PUB01830
      PUB01840
      PUB01850
      PUB01860

```

```

105 FORMAT('OEST #',I1,' => ',(T11,8(1X,E13.6,1X))) )
12 CONTINUE
   NSEED=IX
   RETURN
END
SUBROUTINE EST(AMDA,NUM,T,NOREP)
DIMENSION AMDA(NUM,1),T(NOREP,1)
DOUBLE PRECISION SUM,SUMI
DUMV=ALOG(100.0)
DO 1 K=1,NUM
  GET TBARS
  SUM=0.0D+00
  SUMI=0.0D+00
DO 2 J=1,NOREP
  SUM=SUM+T(J,K)
  IT=T(J,K)
  SUMI=SUMI+IT
CONTINUE
C GET MAXIMUM LIKELIHOOD ESTIMATORS
V=DUMV
IF(SUM.GT.0.0D+00)V=NOREP/SUM
AMDA(K,2)=V
AMDA(K,4)=V
V=DUMV
IF(SUMI.GT.0.0D+00)V=NOREP/SUMI
VI=1.0D+00+V
VI=ALOG(VI)
AMDA(K,3)=VI
AMDA(K,5)=VI
C GET RESTRICTED MLE IF NECESSARY
KM1=K-1
DO 3 I=4,5
  IM2=I-2
DO 4 L=1,KM1
  LL=K-L
  IF(AMDA(LL,I).GE.AMDA(K,I))GO TO 5
CONTINUE
  LL=0
CONTINUE
  LL=LL+1
  IF(LL.GE.K)GO TO 3
SUM=0.0D+00
DO 6 L=LL,K
  TEST=AMDA(L,IM2)
  IF(TEST.LT.0.0)TEST=DUMV
SUM=SUM+1.0D+00/TEST
CONTINUE
  LEN=K-LL+1
6

```

PUB01870  
PUB01880

PUB01890  
PUB01900  
PUB01910  
PUB01920  
PUB01930

PUB01940  
PUB01950  
PUB01960  
PUB01970  
PUB01980  
PUB01990  
PUB02000  
PUB02010  
PUB02020  
PUB02030

PUB02050  
PUB02060

PUB02080  
PUB02090  
PUB02100  
PUB02110  
PUB02120  
PUB02130  
PUB02140  
PUB02150  
PUB02160  
PUB02170  
PUB02180  
PUB02190  
PUB02200  
PUB02210  
PUB02220  
PUB02230  
PUB02240

PUB02260  
PUB02270

```

V=LEN/SUM
IF(I.EQ.5)V=ALOG(1.0+V)
DO 7 L=LL,K
  AMDA(L,I)=V
CONTINUE
C
3
  ATTRIBUTE DATA
SUMI=0.0D+00
DO 8 J=1,NOREP
  IT=T(J,K)
  IF(IT.GE.1)SUMI=SUMI+1.0D+00
CONTINUE
  IF(SUMI.EQ.0.0D+00)SUMI=.01D+00
  IF(SUMI.GE.NOREP)SUMI=SUMI-1.0D-02
  PS=SUMI/NOREP
  AMDA(K,6)=-ALOG(PS)
CONTINUE
  RETURN
END
SUBROUTINE FINISH(NOUT)
  CALL DMPOUT(NOUT)
  WRITE(6,100)
STOP
100
  FORMAT('1 PROGRAM TERMINATING DO TO "FINISH" COMMAND OR "/*"' )
END
SUBROUTINE WOLMAN(AMDA,NO,B,FITTED)
  EXTERNAL RQF
  COMMON/EXPTIM/NOR,T(2500)
  COMMON/WORK/W(600)
  DIMENSION FITTED(NO,1)
  REAL*8 R,Q
  DIMENSION AMDA(NO,1),X(50),Y(50)
  EQUIVALENCE (X(1),R),(X(3),Q)
  DUM=ALOG(1.0E+30)
  N=NO
  B1=B
  WRITE(6,201)B1
  FORMAT('1',WOLMAN MODEL ESTIMATES, LAMDA(I) = -LN(1-R-Q*(',
1E13.6,'))*(1-1)),',',EST #2 THROUGH #6 ARE FROM REGRESSION',
2, ' ON LAMDA ESTIMATES #2 THROUGH #6,',/,
3, ' EST #1 IS A MLE BASED ON LAMDA ESTIMATE #3',/)
  KSTART=2
REGRESSION
C
  DO 1 K=KSTART,6
  DO 2 L=1,N
    Y(L)=1.0-EXP(-AMDA(L,K))
    X(L)=B1**((L-1)
2
  CONTINUE

```

```

PUB02280
PUB02290
PUB02300
PUB02310
PUB02320
PUB02330
PUB02340
PUB02350
PUB02360
PUB02370
PUB02380
PUB02390
PUB02400
PUB02410
PUB02420
PUB02430
PUB02440
PUB02450
PUB02460
PUB02470
PUB02480
PUB02490
PUB02500
PUB02510
PUB02520
PUB02530
PUB02540
PUB02550
PUB02560
PUB02570
PUB02580
PUB02590
PUB02600
PUB02610
PUB02620
PUB02630
PUB02640
PUB02650
PUB02660
PUB02670
PUB02680
PUB02690
PUB02700
PUB02710
PUB02720

```

PUB02730  
PUB02740  
PUB02750

PUB02770  
PUB02780  
PUB02790  
PUB02800  
PUB02810  
PUB02820  
PUB02830  
PUB02840  
PUB02850  
PUB02860  
PUB02870  
PUB02880

PUB02900  
PUB02910  
PUB02920  
PUB02930  
PUB02940  
PUB02950  
PUB02960  
PUB02970  
PUB02980  
PUB02990  
PUB03000  
PUB03010  
PUB03020  
PUB03030  
PUB03040  
PUB03050  
PUB03060  
PUB03070  
PUB03080  
PUB03090  
PUB03100

```

W(600)=5.0
CALL FIT(Y,X,RR,QQ,N)
DO 3 L=1,N
  PONENT=1.0-RR-QQ*X(L)
  IF(PONENT.LE.1.E-30)GO TO 100
  FITTED(L,K)=-ALOG(PONENT)
GO TO 3
CONTINUE
FITTED(L,K)=DUM
CONTINUE
R=RR
Q=QQ
WRITE(6,202)R,Q,K,(FITTED(L,K),L=1,N)
FORMAT(10R,E13.6,10R,E13.6,
1/,1EST #,11,1=>,111,8(1X,E13.6,1X)) )
CONTINUE
C
C MAXIMUM LIKELIHOOD ESTIMATOR BASED ON ATTRIBUTE DATA
C=RQFI(BI,N)
CALL ZSYSTEM(RQF,1.0D-04,4,2,X,100,W,IER)
IF(IER.NE.0)GO TO 50
DO 12 L=1,N
  PONENT=1.0-X(1)-X(3)*BI**(L-1)
  IF(PONENT.LT.1.0E-30)GO TO 101
  FITTED(L,1)=-ALOG(PONENT)
GO TO 12
CONTINUE
FITTED(L,1)=DUM
CONTINUE
K=1
WRITE(6,202)R,Q,K,(FITTED(L,K),L=1,N)
GO TO 11
CONTINUE
DO 4 L=1,N
  FITTED(L,1)=0.0
CONTINUE
RETURN
END
SUBROUTINE FIT(Y,X,A,B,N)
COMMON/WORK/YY(600)
DIMENSION Y(1),X(1)
DOUBLE PRECISION SUMX,SUMY,SUMXY,SUMXX
DATA IENTER/0/
IENTER=IENTER+1
M=YY(600)+.5
NN=N
SUMX=0.0
SUMY=0.0

```



PUB03110  
PUB03120  
PUB03150  
PUB03170  
PUB03180  
PUB03190  
PUB03200  
PUB03210  
PUB03220  
PUB03230  
PUB03240  
PUB03250  
PUB03260  
PUB03270  
PUB03280  
PUB03290  
PUB03300  
PUB03310  
PUB03330  
PUB03350  
PUB03360  
PUB03370  
PUB03380  
PUB03390  
PUB03400  
PUB03410  
PUB03420  
PUB03430  
PUB03440  
PUB03450  
PUB03460  
PUB03470  
PUB03480  
PUB03490  
PUB03500  
PUB03510  
PUB03520  
PUB03530  
PUB03540  
PUB03550  
PUB03560  
PUB03570  
PUB03580  
PUB03590  
PUB03600  
PUB03610  
PUB03620  
PUB03630

```

SUMXY=0.0
SUMXX=0.0
FORMAT(:1PLGT OF DATA TO BE FITTED',//)
CONTINUE
DO 1 K=1,N
SUMX=SUMX+X(K)
SUMY=SUMY+Y(K)
SUMXY=SUMXY+X(K)*Y(K)
SUMXX=SUMXX+X(K)*X(K)
CONTINUE
XBAR=SUMX/N
BNUM=SUMXY-XBAR*SUMY
BDEN=SUMXX-XBAR*SUMX
B=BNUM/BDEN
A=SUMY/N-B*XBAR
DO 99 K=1,N
YY(K)=A+B*X(K)
CONTINUE
IF(IENTER.EQ.M) IENTER=0
RETURN
END
FUNCTION RQFI*8(B,N)
REAL*8 SUM,BB,X(2),TT(50)
REAL*8 R,Q
REAL*8 RQF
COMMON/EXPTIM/NOR,T(2500)
BB=B
NN=N
INDX=0
DO 1 L=1,NN
SUM=0.0D+00
DO 4 K=1,NOR
INDX=INDX+1
IT=TT(INDX)
SUM=SUM+IT
CONTINUE
TT(L)=SUM/NOR
CONTINUE
RQFI=NOR
RETURN
ENTRY RQF(X,K)
SUM=0.0D+00
R=X(1)
Q=X(2)
GO TO (10,20),K
CONTINUE
DO 2 L=1,NN
SUM=SUM+IT(L)/(1.0D+00-R-Q*BB**(L-1))

```



PUB03640  
PUB03650  
PUB03660  
PUB03670  
PUB03680  
PUB03690  
PUB03700  
PUB03710  
PUB03720  
PUB03730  
PUB03740  
PUB03750

```

2      SUM=SUM-1.0D+00/(R+Q*BB**(L-1))
      CONTINUE
      RQF=SUM
      RETURN
20     CONTINUE
      DO 3 L=1,N
      SUM=SUM+T(L)/(1.0D+00-R-Q*BB**(L-1))*BB**(L-1)
      SUM=SUM-1.0D+00/(R+Q*BB**(L-1))*BB**(L-1)
      CONTINUE
      RQF=SUM
      RETURN
      END
      SUBROUTINE GE(AMDA,NQ,FITTED)
      DIMENSION AMDA(NQ,1),FITTED(NQ,1),X(50),Y(50)
      COMMON/WORK/W(600)
      COMMON/EXPTIM/NOR,T(2500)
      REAL*8 CUMT,SUMREP
      WRITE(6,201)
      FORMAT(//,' GENERAL ELECTRIC MODEL ESTIMATES, LAMDA(T) = ',
1,C*T*(-ALPHA),',//',' ALL ESTIMATES ARE FROM REGRESSIONS ON ',
2, ' CORRESPONDING LAMDA ESTIMATES',//)
      W(600)=6.0
      DO 3 NOEST=1,6
      CUMT=0.0D+00
      INDX=0
      DO 1 K=1,NQ
      SUMREP=0.0D+00
      DO 2 J=1,NOR
      INDX=INDX+1
      TI=T(INDX)
      IT=TI+1
      GO TO (10,10,20,10,20,30),NOEST
10     CONTINUE
      SUMREP=SUMREP+TI
      GO TO 2
20     CONTINUE
      SUMREP=SUMREP+IT
      GO TO 2
30     CONTINUE
      SUMREP=SUMREP+NOR
      J=NOR
      CONTINUE
      CUMT=CUMT+SUMREP
      XK=CUMT
      X(K)=-ALOG(XK)
      IF(AMDA(K,NOEST).LE.1.0E-60)GO TO 300
      Y(K)=ALOG(AMDA(K,NOEST))
      CONTINUE
1

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```

CALL FIT(Y,X,AK,ALPHA,NO)
AK=EXP(AK)
DO 4 K=1,NO
  FITTED(K,NOEST)=AK*EXP(X(K)*ALPHA)
CONTINUE
202 WRITE(6,202)AK,ALPHA,NOEST,(FITTED(L,NOEST),L=1,NO)
FORMAT(10A=1,E13.6,1,ALPHA=1,E13.6,
1/,1,EST #1,1,1=>1,(T11,8(1X,E13.6,1X)))
GO TO 3
300 WRITE(6,203)NOEST,K,AMDA(K,NOEST)
203 FUMAT(10ERRR IN EST #1,11,/,1,VALUE FOR FAILURE NO 1,
112,1,IS 1,E13.6)
DO 31 K=1,NO
  FITTED(K,NOEST)=0.0
CONTINUE
CONTINUE
RETURN
END
SUBROUTINE WEISS(AMDA,NO,FITTED)
EXTERNAL DBIFDI
REAL*8 BB
DIMENSION AMDA(NO,1),FITTED(NO,1),X(50),Y(50)
COMMON/WORK/W(600)
W(600)=5.0
WRITE(6,201)
FORMAT(//,1,WEISS MODEL ESTIMATES, LAMDA(I) = 1,
1,1.0/A + B/(A*1),/,1,EST #2 THROUGH #6 ARE FROM REGRESSION',
2,ON LAMDA ESTIMATES #2 THROUGH #6,/,/,
3,EST #1 IS A MLE BASED ON LAMDA ESTIMATE #2,/)
DO 2 NOEST=2,6
  DO 1 K=1,NO
    Y(K)=AMDA(K,NOEST)
    X(K)=1.0/K
  CONTINUE
  CALL FIT(Y,X,A,B,NO)
  DO 3 K=1,NO
    FITTED(K,NOEST)=A+B*X(K)
  CONTINUE
  A=1.0/A
  B=B*A
202 WRITE(6,202)A,B,NOEST,(FITTED(L,NOEST),L=1,NO)
FORMAT(10A=1,E13.6,1,B=1,E13.6,/,
1,EST #1,1,1=>1,(T11,8(1X,E13.6,1X)))
CONTINUE
C=BF1(NO)
B=0.0
LOOPS=50
DO 10 I=1,LOOPS

```

```

10 DELTB=BIFDI(B)/DERIV(B)
11 B=B-DELTB
12 IF(ABS(DELTB/B).LT.1.0E-04)GO TO 11
4 CONTINUE
  GO TO 12
  CONTINUE
  A=HAY(B)
  DO 4 K=1,NO
    FITTED(K,1)=1.0/A+B/A*X(K)
  CONTINUE
  NOEST=1
  WRITE(6,202)A,B,NOEST,(FITTED(L,NOEST),L=1,NO)
  RETURN
12 CONTINUE
  BB=B
  IOO=100
  CALL ZREAL1(DBIFDI,1.D-04,0.D+00,0.D+00,4,1,BB,IOO,IER)
  B=BB
  IF(IER.EQ.0)GO TO 11
  WRITE(6,204)
  FORMAT('OERROR IN ESTIMATE #6. SOLUTION TO B FAILED TO CONVERGE.')
```

204

```

  DO 31 K=1,NO
    FITTED(K,NOEST)=0.0
  CONTINUE
  RETURN
31
```

```

END
FUNCTION BFI(NO)
COMMON/EXPTIM/NJR,T(2500)
REAL*8 SUM,SUMT,SUMTDI
REAL*8 SUMI
SUMT=0.0D+00
SUMTDI=SUMT
INDX=0
DO 2 K=1,NO
  SUM=0.0D+00
  DO 1 J=1,NJR
    INDX=INDX+1
    SUM=SUM+T(INDX)
  CONTINUE
  SUM=SUM/NJR
  SUMT=SUMT+SUM
  SUMTDI=SUMTDI+SUM/K
  CONTINUE
  BFI=SUMT/SUMTDI
  RETURN
ENTRY BIFDI(B)
SUM=0.0D+00
SUMI=0.0D+00
```

1

2

```

3      DO 3 K=1,NO
        SUM=SUM+1.0/(B+K)
        SUMI=SUMI+K/(B+K)
        CONTINUE
        BIFDI=SUMT*SUM-SUMTDI*SUMI
        RETURN
        ENTRY DERIV(B)
        SUM=0.0D+00
        SUMI=0.0D+00
        DO 4 K=1,NO
            SUM=SUM+1.0/(B+K)**2
            SUMI=SUMI+K/(B+K)**2
            CONTINUE
            DERIV=SUMTDI*SUMI-SUMT*SUM
            RETURN
            ENTRY HAY(B)
            HAY=(B*SUMTDI+SUMT)/NO
            RETURN
        END
        REAL FUNCTION DBIFDI*8(BB)
        REAL*8 BB
        B=BB
        DBIFDI=BIFDI(B)
        RETURN
    END
    SUBROUTINE WOODS(AMDA,NO,FITTED)
    DIMENSION AMDA(NO,1),FITTED(NO,1),X(50),Y(50)
    COMMON/WORK/W(600)
    COMMON/EXPTIM/NOK,T(2500)
    TEST=ALOG(.999999)
    WRITE(6,201)
    FORMAT(//,' WOODS-CHERNOFF MODEL ESTIMATES, LAMDA(I) = -',
1, LN(1.0 - EXP(-ALPHA - BETA*(I-1))),',', ' ALL ESTIMATES',
2, ' ARE FROM REGRESSIONS ON CORRESPONDING LAMDA ESTIMATES',//)
    W(600)=6.0
    DO 3 NOEST=1,6
        DO 1 K=1,NO
            X(K)=K-1
            IF(AMDA(K,NOEST).LE.-TEST)GO TO 300
            Y(K)=-ALOG(1.0-EXP(-AMDA(K,NOEST)))
            CONTINUE
            CALL FIT(Y,X,ALPHA,BETA,NO)
            DO 2 K=1,NO
                ONENT=-(ALPHA+BETA*X(K))
                IF(ONENT.GE.TEST)ONENT=TEST
                FITTED(K,NOEST)=-ALOG(1.0-EXP(ONENT))
            CONTINUE
            WRITE(6,202)ALPHA,BETA,NOEST,(FITTED(L,NOEST),L=1,NO)

```

```

202 FORMAT('OALPHA = ',E13.6,' ',BETA = ',E13.6,/,
10 EST #',I1,' => ',(I11,8(I1X,E13.6,1X)))
GO TO 3
300 WRITE(6,203)NOEST,K,AMDA(K,NOEST)
203 FORMAT('OERRCR IN EST #',I1,/, ' VALUE FOR FAILURE NO ',
112,' IS ',E13.6)
DO 31 K=1,NO
FITTED(K,NOEST)=0.0
CONTINUE
CONTINUE
RETURN
END
SUBROUTINE UERTST(NAME)
LOGICAL*1 NAME(6)
RETURN
END
SUBROUTINE STORE(NMEAS,VALUES,TITLE)
DIMENSION TITLE(1),VALUES(8,1),OUTPUT(10,60)
DATA NCOL/0/
NCOL=NCOL+1
GO TO 1
ENTRY RESET
NCOL=0
RETURN
CONTINUE
DO 2 K=1,3
OUTPUT(K,NCOL)=TITLE(K)
CONTINUE
OUTPUT(4,NCOL)=NMEAS
DO 3 K=5,10
KM4=K-4
OUTPUT(K,NCOL)=VALUES(NMEAS,KM4)
CONTINUE
RETURN
ENTRY DMPOUT(NOUT)
IF(NCOL.EQ.0)RETURN
WRITE(NOUT,101)
FORMAT(1,/,/, ' COMPARISON OF MODELS ',
1//, ' MODEL ',6(' ',E13.6,1X),/,
2//, ' MEASURE ',6(' ',E13.6,1X),/,
DO 5 K=1,NCOL
WRITE(NOUT,102)(OUTPUT(L,K),L=1,10)
FORMAT(1,/,3A4,' ',F6.0,2X,6(' ',E13.6))
CONTINUE
WRITE(NOUT,101)(L,L=2,22,4)
DO 6 K=1,NCOL
ITEST=(K+1)/4*4-(K+1)
IF(ITEST.NE.0) WRITE(NOUT,102) OUTPUT(1,K),OUTPUT(2,K),

```



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C C C C C C C C C C
PROGRAM TO GENERATE MODEL I DATA PROGRAMMED ON HP 9810
DEFINITION OF MODEL IS A FIXED POINT EQUATION
WHICH WAS SOLVED RECURSIVLY

PROGRAM TO GENERATE MODEL II DATA PROGRAMMED ON HP 9810
GENERATED IN STRAIGHT FORWARD MANNER FROM DEFINITION

PROGRAM TO GENERATE MODEL III DATA

REAL*4 LAMDA(16)
ALPHA=-ALOG(1.0-EXP(-.7))
BETA=ALOG((1.0-EXP(-.7))/(1.0-EXP(-.05)))
BETA=BETA/15
DO 1 K=1,16
  KMI=K-1
  LAMDA(K)=-ALOG(1.0-EXP(-ALPHA-BETA*KMI))
CONTINUE
WRITE(6,101)ALPHA,BETA,(LAMDA(I),I=1,16)
FORMAT(10ALPHA=,E15.7,; BETA=,E15.7,
1/, (5E18.9,/))
WRITE(7,102)LAMDA
FORMAT(5F10.9)
STOP
END
1
101
102

C C C
PROGRAM TO GENERATE MODEL IV DATA

REAL*4 LAMDA(16)
B=.7
RPQ=1.0-EXP(-.7)
RPQ15=1.0-EXP(-.05)
Q15MQ=RPQ15-RPQ
Q=Q15MQ/(8*15-1.0)
R=RPQ-Q
DO 1 K=1,16
  KMI=K-1
  LAMDA(K)=-ALOG(1.0-R-Q*B**KMI)
CONTINUE
WRITE(6,101)R,Q,B
FORMAT(10LAMDA(I)=-ALOG(1.0-(,E15.7,)-(,
1E15.7,)*(,E15.7,)**(1-1),)
WRITE(7,102)( LAMDA(I),I=1,16)
FORMAT(5F10.9)
STOP
END
1
101
102

```



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